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REPORT DOCUMENTATION PAGE	READ INSTRUCTIONS BEFORE COMPLETING FORM	
AFOSR-TK- 87-1015	J. RECIPIENT'S CATALOG NUMBER	
4. TITLE (and Subtitle)	5. TYPE OF REPORT & PERIOD COVERED	
On the least squares estimator in moving average		
models of order one	6. PERFORMING ONG, REPORT NUMBER 86-45	
7. AUTHOR(s)	.B. CONTRACT OR GRANT NUMBER(+)	
H.A. Niroomand Chapeh and M. Bhaskara Rao	N00014-85-K-0292 F49620-85-C-0008	
8. PERFORMING ORGANIZATION NAME AND ADDRESS	10. PROGRAM ELEMENT, PROJECT, TASK AREA & WORK UNIT NUMBERS	
Center for Multivariate Analysis University of Pittsburgh, 515 Thackeray Hall	6162F	
Pittsburgh, PA 15260	23014 A5	
II AND	12. REPORT DATE	
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14. MONITORING ACENCY NAME & ADDRESS(II different from Controlling Office)	18. SECURITY CLASS, (of this report)	
< 0 0 0 0 11	unclassified	
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16. DISTHIBUTION STATEMENT (of this Report)		
Approved for public-release; distribution unlimited.		
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18. SUPPLEMENTARY NOTES	OCT 1 0 1967 1	
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Consistency, Least squares estimator, Moving average	ge model.	

20 ABSTRACT (Cinitinue on reverse side if necessary and identity by block number)

A simple expression is derived in this paper for the error sum of squares in the context of moving average models of order one. A computer program is developed to estimate the parameter of a moving average model of order one based on the method of least squares.

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Technical Report No. 86-45

December 1986

Center for Multivariate Analysis Fifth Floor, Thackeray Hall University of Pittsburgh Pittsburgh, PA 15260

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*This work is supported by Contract N00014-85-K-0292 of the Office of Naval Research and Contract F49620-85-C-0008 of the Air Force Office of Scientific Research. The United States Government is authorized to reproduct and distribute reprints for governmental purposes notwithstanding any copyright notation heron.

By H.A. Niroomand Chapeh and M. Bhaskara Rao University of Sheffield, UK

Keywords : Moving average model; Least squares estimator; Consistency

Language

Fortran 77

Description and purpose

Given a time series data, estimate β in the moving average model $Y_t = e_t + \beta e_{t-1}$, t = 1,2,3,... of order one using the method of least squares.

Theory

Let $Y_t = e_t + \beta e_{t-1}$, t = 1,2,3,... be a moving average process of order one, where $e_o = 0$ and e_1 , e_2 , ... is a sequence of independent identically distributed random variables with mean zero and variance σ^2 . There are several methods of estimation are available in the literature for estimating β for a given time series data $y_1, y_2, ..., y_N$. McClave (1974) compared the performance of some five estimators of β for moderate sample sizes by simulating the process Y_t , t = 1,2,...,100. All these methods use approximations of one kind or the other and involve selection of some indices for a good degree of approximation. Consequently, the execution of these methods looks complex and it is natural co search for a simple method which performs competitively well with these methods. The method of least square is intuitively very appealing, and recently, under some conditions, Macpherson and Fuller (1983) showed that the least squares

estimator is consistent for ß in [-1,1]. We have compared the performance of least squares estimator for moderate sample sizes vis-a-vis with the five estimators examined by McClave (1974). See Niroomand Chapeh and Bhaskara Rao (1987). The least squares estimator comes out better than these five estimators. Additionally, as the present article demonstrates, the execution of the least squares method is much simpler than these five methods.

Following Macpherson and Fuller (1983), the least squares estimator of β is that value of θ in [-1,1] that minimizes

$$Q_{N}(\theta) = \Sigma_{t=1}^{N} [e_{t}(Y:\theta)]^{2},$$

where $e_t(Y:\theta) = Y_t - \theta e_{t-1}(Y:\theta)$, t = 1,2,...,N and $e_0(Y:\theta) = 0$. It works out that

$$= \mathbf{e}_{k}(\mathbf{Y}:\theta) = \mathbf{Y}_{k} - \theta \mathbf{Y}_{k-1} + \theta^{2} \mathbf{Y}_{k-2} - \ldots + (-1)^{k-1} \theta^{k-1} \mathbf{Y}_{1},$$

k = 1,2,...,N. Consequently, $\,Q_{N}^{}(\theta)\,$ is a polynomial in $\,\theta\,$ of degree $\,2N\text{--}2\,.$ More explicitly, for N even,

$$Q_{N}(\theta) = \sum_{i=1}^{N} Y_{i}^{2} + (\sum_{i=1}^{N-1} Y_{i}^{2} + 2 \sum_{i=1}^{N-2} Y_{i}Y_{i+2})\theta^{2} + \dots + (\sum_{i=1}^{N-2} Y_{i}^{2} + 2 \sum_{i=1}^{N-3} Y_{i}Y_{i+2} + 2 \sum_{i=1}^{N-4} Y_{i}Y_{i+4})\theta^{4} + \dots + (\sum_{i=1}^{N/2+1} Y_{i}^{2} + 2 \sum_{i=1}^{N/2} Y_{i}Y_{i+2} + \dots + 2 \sum_{i=1}^{2} Y_{i}Y_{i+N-2})\theta^{N-2} + (\sum_{i=1}^{N/2} Y_{i}^{2} + 2 \sum_{i=1}^{N/2-1} Y_{i}Y_{i+2} + \dots + 2 \sum_{i=1}^{1} Y_{i}Y_{i+N-2})\theta^{N} + (\sum_{i=1}^{N/2-1} Y_{i}^{2} + 2 \sum_{i=1}^{N/2-2} Y_{i}Y_{i+2} + \dots + 2 \sum_{i=1}^{1} Y_{i}Y_{i+N-2})\theta^{N} + (\sum_{i=1}^{N/2-1} Y_{i}^{2} + 2 \sum_{i=1}^{N/2-2} Y_{i}Y_{i+2} + \dots + 2 \sum_{i=1}^{1} Y_{i}Y_{i+N-4})\theta^{N+2} + \dots + \sum_{i=1}^{N/2-2} Y_{i}Y_{i+2} + \dots + 2 \sum_{i=1}^{1} Y_{i}Y_{i+N-4})\theta^{N+2} + \dots + \sum_{i=1}^{N/2-2} Y_{i}Y_{i+2} + \dots + 2 \sum_{i=1}^{N/2-2} Y_{i}Y_{i+N-4} + \dots + \sum_{i=1}^{N/2-2} Y_{i}Y_{i+1} + \dots + \sum_{i=1}^{N/2-2} Y_{i}Y_{i+N-4} + \dots + \sum_{i=1}^{N/2-2} Y_{i}Y_{i+N-4} + \dots + \sum_{i=1}^{N/2-2} Y_{i}Y_{i+1} + \dots + \sum_{i=1}^{N$$

$$-2(\Sigma_{i=1}^{N-2} Y_{i}Y_{i+1} + \Sigma_{i=1}^{N-3} Y_{i}Y_{i+3})\theta^{3} - \dots$$

$$-2(\Sigma_{i=1}^{N/2} Y_{i}Y_{i+1} + \Sigma_{i=1}^{N/2-1} Y_{i}Y_{i+3} + \dots + \Sigma_{i=1}^{1} Y_{i}Y_{i+N-1})\theta^{N-1}$$

$$-2(\Sigma_{i=1}^{N/2-1} Y_{i}Y_{i+1} + \Sigma_{i=1}^{N/2-2} Y_{i}Y_{i+3} + \dots + \Sigma_{i=1}^{1} Y_{i}Y_{i+N-3})\theta^{N+1}$$

$$- \dots -2Y_{1}Y_{2}\theta^{2N-3}.$$

The above expression looks formidable to include in a computer program. However, the above can be simplified as follows. Let A be the matrix of order NxN defined by

$$\begin{bmatrix} \mathbf{Y}_1 & \mathbf{0} & \mathbf{0} & \mathbf{0} & \dots & \mathbf{0} & \mathbf{0} \\ \mathbf{Y}_2 & -\mathbf{Y}_1 & \mathbf{0} & \mathbf{0} & \dots & \mathbf{0} & \mathbf{0} \\ \mathbf{Y}_3 & -\mathbf{Y}_2 & \mathbf{Y}_1 & \mathbf{0} & \dots & \mathbf{0} & \mathbf{0} \\ \mathbf{Y}_4 & -\mathbf{Y}_3 & \mathbf{Y}_2 & -\mathbf{Y}_1 & \dots & \mathbf{0} & \mathbf{0} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \mathbf{Y}_{N-1} & -\mathbf{Y}_{N-2} & \mathbf{Y}_{N-3} & -\mathbf{Y}_{N-4} & & (-1)^{N-2}\mathbf{Y}_1 & \mathbf{0} \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ \mathbf{Y}_N & -\mathbf{Y}_{N-1} & \mathbf{Y}_{N-2} & -\mathbf{Y}_{N-3} & \dots & (-1)^{N-2}\mathbf{Y}_2 & (-1)^{N-1}\mathbf{Y}_1 \end{bmatrix} \vdots$$

Let 9 be the column vector of order Nxl defined by

$$\theta^{T} = [1, \theta, \theta^{2}, \theta^{3}, \dots, \theta^{N-1}],$$

where T stands for operation transpose. It can be verified that

$$Q_{N}(\theta) = \theta^{T}(A^{T}A)\theta.$$

We use this simple form of $Q_{\mathbf{u}}(\theta)$ in the computer program to estimate 3.

A case study

We used the method of least squares to fit a first order moving average model to the first order differences of the time series data "IBM Common Stock Closing Prices - Daily, 17th May, 1961 to 2nd November, 1962". See Box and Jenkins (1976, p.239 and p.526). The fitted model works out to be

$$\nabla Y_{t} = e_{t} + 0.089e_{t-1}$$

 $\nabla Y_t = Y_t - Y_{t-1}$ and Y_i 's are the original series, Box and Jenkins obtained the estimate of β to be 0.09.

Structure

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SUBROUTINE ZXMWD(FCN, M, NSEG, Al, Bl, NSRCH, X, F, WORK, IWORK, IER)

Formal parameters

M __ Integer Input: number of unknowns parameters

NSIG Integer Input: number of digits of accuracy required in the parameter estimates

Al, Bl Real arrays Constraint vectors of length of M

Input: X(I) is required to satisfy - Al(I).LE.X(I).LE.Bl(I)

NSRCH Integer Number of starting points to be generated

Input: suggested value = min(2**M+5,100)

X Real array Vector of length M containing the final parameter estimates (output)

F Real Value of the function at the final parameter estimates (output)

WORK Real array Work vector of length M*(M+1)/2+11*M

IWORK Integer array

Work vector of length M

IER Real Error parameter (output)

Terminal error

IER=130 indicates that it was not possible to calculate
 the solution to NSIG digits (See remarks)

IER=132 indicates that Al(I).GE.Bl(I) for some I = 1, 2, ..., N. No attempt is made to find the minimum in this case.

Remarks

When IER is returned as 130 or 131, the parameter estimates in X may not be reliable. Futher checking should be performed. Use of a larger NSRCH may produce more reliable parameter estimates.

Auxiliary algorithms

SUBROUTINE ZXMWD will use the FCN(M,X,F) to minimize F

M Integer Input: number of unknown parameters

X Real array Vector of length M containing the final parameter estimates (output)

F Real Output: Value of the function at the final parameter estimates

SUBROUTINE VMULFF(AMATT,AMAT,L,KA,KB,IA,IB,C,IC,IER) will find the multiplication of two matrices AMATT and AMAT

Formal parameters

AMATT Real array Input: N by N matrix stored in full storage mode

AMAT Real array Input: N by N matrix stored in full storage mode

L Integer Input: number of rows in AMATT

KA Integer Input: number of columns in AMATT

KB Integer Input: number of columns in AMAT

IA Integer Input: row dimension of matrix AMATT

IB Integer Input: row dimension of matrix AMAT

C Real array Output: N by N matrix containing the product C = AMATT*AMAT

IC Integer Input: row dimension of matrix C

IER Integer Output: Error indicator

IER=129 indicates that AMATT or AMAT or C was dimensioned

incorrectly

References

Box, G.E.P. and Jenkins, G.M. (1976) Time Series Analysis: Forecasting and Control.

Holden-Day, Oakland, California

Macpherson, B.D. and Fuller, W.A. (1983) - Consistency of the least squares estimator of the first order moving average parameter, Ann.Stat., 11,326-329.

McClave, J.T. (1974) - A comparison of moving average estimation procedures, Comm.Stat., 3, 865-883.

Niroomand Chapeh, H.A. and Bhaskara Rao, M. (1987) - A comparison of various methods of estimation in moving average models (in preparation).

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THIS PROGRAM WILL FIT THE FIRST DRDER MOVING AVERAGE MODEL TO DATA
           IER,N,L.KA,KB,IA,IB,IC,NSIG,NSRCH,IWORK | M,IV N=
INTEGER
        X(1),A1,B1,WORK(12),F,AMAT(N=,N=,AMATT(N=,N=,,0)N=,0=
 REAL
 EXTERNAL
           FCN
 COMMON
          N.C
 F1=0.0
 F2=0.0
OPEN(UNIT=100, FILE='DATA.DAT', STATUS='OLD')
 OPEN(UNIT=99, FILE='RES.DAT', STATUS='NEW')
 READ(100,*) IV
 M=1
 NSIG=4
 NSRCH=7
Al=1.0
 E1=1.0
 N=
 L=N
 KA=N
 KB=N
 IA=N
 IB=N
 IC=N
 THE GIVEN DATA IS PUT INTO THE MATRIX A AS EXPLAINED IN THE THEORY
            J=1,N
 DO
 DO
            I=J,N
```

```
CONTINUE
     50 1 1=1,3
    J=1, N
     WATT 1,7 = WAT 1,I
     0.187137E
     DAD MOREF AMATT, AMAT, L, KA, KB, IA, IB, C, IC, IER)
     DAIL DOWN FON, M, MSID, A1, B1, NSRCH, M, F, WORK, IWORK, IER)
     WRITE 39,101 X 1.
    FIRMAT NUFINAL RES IS :- AHAT', F10.4)
    5772
     ENT
     SUBROUTINE .FOR M, M, F
 TATECER MATALAN
     PEAR X.11,F,C N=,N=
     COMMON N.C
     F1=0.3
     ±0,3-1 5=0,3-1
     31M1=..1
     De 20 I=1,5+1
20 SUM1=SUM1+0 I,J+2-I
200 F1#F1+(X 11**J)*SUC
      F2=0.0
      DO 360 J=1*N-1,N,-1
```

SUM2=0.0

DO 30 I=J+2-N,N

30 SUM2=SUM2+C(I,J+2-I)

300 F2=F2+(X(1)**J)*SUM2

F=F1+F2

RETURN

END

Concluding remarks: This work is useful for drawing inferences for signal process models when the observations form a moving average model of order one. A manuscript is under preparation detailing applications of the results of this paper to signal processing.

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86-45		
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On the least squares estimator in moving average		technical - December 1986
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H.A. Niroomand Chapeh and M. Bhaska	ra Rao	N00014-85-K-0292 F49620-85-C-0008
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Center for Multivariate Analysis University of Pittsburgh, 515 Thack Pittsburgh, PA 15260	eray Hall	IS. PROGRAM ELEMENT PROJECT TASK AREA & WORK UNIT HUMBERS
11. CONTROLLING OFFICE NAME AND ADDRESS		12. REPORT DATE
Office of Naval Research		December 1986
Air Force Office of Scientific Rese	earch	13. NUMBER OF PAGES
14. MONITORING AGENCY NAME & ADDRESSII dillereni	Iron Controlling Office)	18. SECURITY CLASS. (of this report)
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		ISA. DECLASSIFICATION/DOWNGRADING
16. DISTHIBUTION STATEMENT (of this Report)		
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approved for public release, discri	.Ducion uniimited	•
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17. DISTRIBUTION STATEMENT (of the obstract entered in Bluck 20, if different from Report)		
18. SUPPLEMENTARY NOTES		
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19 KEY WORDS (Continue un reverse side il necessory and identify by block number)		
Consistency, Least squares estimator, Moving average model.		
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